# Linear Algebra

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 $Fall,\ 2021$ 

### Review: Inner products on real linear space

An inner product on V is a function  $\langle , \rangle : V \times V \to \mathbb{R}$  such that

- $\langle v, v \rangle = 0$  if and only if v = 0.

#### Review

• Suppose that V is an inner product space. For  $v \in V$ , we define the norm of v, denoted ||v||, by  $||v|| = \sqrt{\langle v, v \rangle}$ .

• Two vectors  $u, v \in V$  are said to be orthogonal if  $\langle u, v \rangle = 0$ .

### Review: Orthogonal Subspaces

#### Definition

Two subspaces  $W_1$  and  $W_2$  of the same space V are orthogonal, denoted by  $W_1 \perp W_2$ , if and only if each vector  $w_1 \in W_1$  is orthogonal to each vector  $w_2 \in W_2$ :

$$\langle w_1, w_2 \rangle = 0.$$

for all  $w_1$  and  $w_2$  in  $W_1$  and  $W_2$ , respectively.

### Review: Orthogonal complement of a subspace

#### Definition

Given a subspace W in linear space V, the space of all vectors orthogonal to W is called the orthogonal complement of V. It is denoted by  $W^{\perp}$ .

- We emphasize that  $W_1$  and  $W_2$  can be orthogonal without being complements.
- $W_1 = \operatorname{span}((1,0,0))$  and  $W_2 = \operatorname{span}((0,1,0))$ .

#### Fundamental theorem of orthogonality

Review: Fundamental theorem of orthogonality

Let  $A \in M_{mn}(\mathbb{R})$ .

- **①** The row space is orthogonal to the nullspace (in  $\mathbb{R}^n$ ).
- ② The column space is orthogonal to the left nullspace (in  $\mathbb{R}^m$ ).

### Review: Fundamental theorem of orthogonality

Let  $A \in M_{mn}(\mathbb{R})$ .

- **①** The nullspace is the orthogonal complement of the row space in  $\mathbb{R}^n$ .
- ② The left nullspace is the orthogonal complement of the column space in  $\mathbb{R}^m$ .

- From the row space to the column space, A is actually invertible. Every vector in the column space comes from exactly one vector in the row space.

#### Matrix Representation of Inner Products

- Let  $B = \{v_1, \dots, v_n\}$  be a basis for linear space V.
- Suppose that a bilinear function  $\langle \, , \, \rangle : V \times V \to \mathbb{R}$  is an inner product for  $\mathbb{R}^n$ .
- We want to investigate a matrix representation of this inner product.

#### Matrix Representation of Inner Products

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- We want to investigate a matrix representation of this inner product.
- Orthonormal basis!
- Vectors  $q_1, \ldots, q_n$  are orthonormal if

$$q_i^T q_j = \begin{cases} 0 & \text{whenever} & i \neq j \\ \\ 1 & \text{whenever} & i = j \end{cases}$$
 (for orthogonality)

### Change of basis matrix for inner product space

Suppose that  $B = \{v_1, \dots, v_n\}$  and  $B' = \{v'_1, \dots, v'_n\}$  are two bases for an inner product V. Then for each  $v \in V$ , we have

$$[v]_B = P[v]_{B'}$$

such that

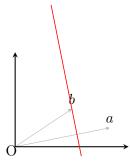
$$v_j' = \sum_{i=1}^n P_{ij} v_r.$$

and P is the change basis matrix.

What is the relationship between the matrix of the inner product relative to the basis B and the basis B'?

#### The Gram-Schmidt Process

- ullet Suppose that a,b are independent vectors, but they are not orthogonal.
- Let  $V = \text{span}(\{a, b\})$ .
- So,  $\{a, b\}$  is a basis for V.
- How can we find a way to make an orthogonal basis?



#### The Gram-Schmidt Process

- ullet Suppose a,b,c are independent but are not orthogonal vectors.
- Let  $V = \text{span}(\{a, b, c\})$ .
- So,  $\{a, b, c\}$  is a basis for V.
- We want to find a way to make an orthogonal basis:

•

$$q_{1} = \frac{1}{\|a\|} a$$

$$q_{2} = \frac{1}{\|b - \langle b, q_{1} \rangle q_{1}\|} (b - \langle b, q_{1} \rangle q_{1})$$

$$q_{3} = \frac{1}{\|c - \langle c, q_{1} \rangle q_{1} - \langle c, q_{2} \rangle q_{2}\|} (c - \langle c, q_{1} \rangle q_{1} - \langle c, q_{2} \rangle q_{2})$$

### Example

$$\bullet \ a = \begin{vmatrix} 1 \\ 0 \\ 1 \end{vmatrix} \quad b = \begin{vmatrix} 1 \\ 0 \\ 0 \end{vmatrix} \quad c = \begin{vmatrix} 2 \\ 1 \\ 0 \end{vmatrix}$$

• The Gram-Schmidt Process:  $q_1 = \frac{1}{\sqrt{2}}a$ 

$$b - \left\langle b, q_1 \right\rangle q_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - \frac{1}{\sqrt{2}} \times \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \quad q_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

$$c - \left\langle c, q_1 \right\rangle q_1 - \left\langle c, q_2 \right\rangle q_2 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} - \sqrt{2} \begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ \frac{1}{\sqrt{2}} \end{bmatrix} - \sqrt{2} \begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ -\frac{1}{\sqrt{2}} \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

#### The Gram-Schmidt process

- The Gram-Schmidt process
  - $\bullet$  starts with independent vectors  $v_1, \ldots, v_n$
  - 2 ends with orthonormal vectors  $q_1, \ldots, q_n$ .
- At step 1:  $q_1 = \frac{1}{\|v_1\|} v_1$ .
- At step j  $(2 \le j \le n)$ :
  - ① it subtracts from  $a_j$  its components in the directions  $q_1, \ldots, q_{j-1}$  that are already settled:

$$Q_j = v_j - \langle v_j, q_1 \rangle q_1 - \dots - \langle v_j, q_{j-1} \rangle q_{j-1}.$$

- **3**  $\operatorname{span}(\{v_1, \dots, v_j\}) = \operatorname{span}(\{q_1, \dots, q_j\}).$

## Thank You!